# Package: localFDA (via r-universe)

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Type Package

Title Localization Processes for Functional Data Analysis

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Description Implementation of a theoretically supported alternative to k-nearest neighbors for functional data to solve problems of estimating unobserved segments of a partially observed functional data sample, functional classification and outlier detection. The approximating neighbor curves are piecewise functions built from a functional sample. Instead of a distance on a function space we use a locally defined distance function that satisfies stabilization criteria. The package allows the implementation of the methodology and the replication of the results in Elías, A., Jiménez, R. and Yukich, J. (2020) <arXiv:2007.16059>.

**License** GPL-3 **Encoding** UTF-8

LazyData true

RoxygenNote 7.1.1

URL https://github.com/aefdz/localFDA

BugReports https://github.com/aefdz/localFDA

**Imports** stats, graphics **Depends** R (>= 2.10)

Repository https://aefdz.r-universe.dev

RemoteUrl https://github.com/aefdz/localfda

RemoteRef HEAD

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classificationData

Two groups of Gaussian processes with different mean values

## Description

Two groups of Gaussian processes with different mean values

## Usage

classificationData

## **Format**

A matrix with n = 100 functions by columns and t = 200 evaluation points by row. The first 50 are G1 and second 50 curves are G2 that differs in the mean value.

## References

Elías, Antonio, Jiménez, Raúl and Yukich, Joe (2020). Localization processes for functional data analysis (submitted).

## **Examples**

```
matplot(classificationData, type = "l")
```

exampleData 3

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Functional Gaussian processes.

#### **Description**

Functional Gaussian processes.

#### **Usage**

exampleData

#### **Format**

A matrix with n = 1000 functions by columns and t = 100 evaluation points by row.

#### References

Elías, Antonio, Jiménez, Raúl and Yukich, Joe (2020). Localization processes for functional data analysis (submitted).

#### **Examples**

```
matplot(exampleData, type = "1")
```

localizationClassifier

Localization classifier

#### **Description**

Given a training sample with g groups, it predicts the group of the test sample.

## Usage

```
localizationClassifier(trainingSample, testSample, classNames, k_opt, g_pi)
```

## **Arguments**

trainingSample	matrix p by n,	being n the i	number of func	ctions and p t	the number of grid	points.
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The colnames of the trainingSample matrix are i\_groupName where i goes from

1 to the sample size of the group.

testSample matrix p by n, being n the number of functions to classify and p the number of

grid points.

classNames character vector with the group names.

k\_opt Maximum order of the localization processes used in the classification rule.

g\_pi Vector of size g with a priori probabilities for the bayes classifier. If it is missing

the probability is defined by the proportion of curves of each group.

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#### Value

Two named training and test. Training contains the estimations made with the training sample (localization statistics and localization distances). Test contains the classification results (for each incoming data, localization distances in each group, prior probabilities used, likelihood in each group and the predicted\_class).

#### References

Elías, Antonio, Jiménez, Raúl and Yukich, Joe (2020). Localization processes for functional data analysis (submitted).

#### **Examples**

```
X <- classificationData
ids_training <- sample(colnames(X), 90)
ids_testing <- setdiff(colnames(X), ids_training)
trainingSample <- X[,ids_training]
testSample <- X[,ids_testing]; colnames(testSample) <- NULL #blind
classNames <- c("G1", "G2")
classification_results <- localizationClassifier(trainingSample, testSample, classNames, k_opt = 3)</pre>
```

localizationDistances Localization distances

## **Description**

Compute the localization distances of order k of the curve y0.

#### Usage

```
localizationDistances(y, y0)
```

## Arguments

y matrix p by n, being n the number of functions and p the number of grid points. y0 focal curve (index or character name).

#### Value

a vector of length (n-1), being the localization distance of its corresponding order.

#### References

Elías, Antonio, Jiménez, Raúl and Yukich, Joe (2020). Localization processes for functional data analysis (submitted).

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## **Examples**

 $localization \texttt{Distances\_1} \ \texttt{'-localizationDistances} (example \texttt{Data}, \ y \emptyset \ \texttt{= "1"})$ 

localizationProcesses Localization processes

## **Description**

Compute the localization processes of order k of the curve y0.

## Usage

```
localizationProcesses(y, y0)
```

## **Arguments**

y matrix p by n, being n the number of functions and p the number of grid points.

y0 focal curve index or name

## Value

a list with one element, 1c, a matrix of size p x (n-1), being the (n-1) columns the localization processes of its corresponding order.

#### References

Elías, Antonio, Jiménez, Raúl and Yukich, Joe (2020). Localization processes for functional data analysis (submitted).

# Examples

```
localizationProcesses_1 <- localizationProcesses(exampleData, y0 = "1")</pre>
```

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localizationStatistics

Localization Distances Statistics

## **Description**

Estimate the mean and standard deviation of the localization distances mean.

#### **Usage**

```
localizationStatistics(y, robustify = TRUE, whiskerrule)
```

#### **Arguments**

y matrix p by n, being n the number of functions and p the number of grid points.

robustify if TRUE the mean and standard deviation are estimated with a the trimmed sam-

ple. Default is TRUE.

whiskerrule Range parameter for the univariate boxplot detection rule. Default = 3.

#### Value

a list with the localization distances of each function (localizationDistances), the estimated mean (mean) and standard deviation (sd).

## References

Elías, Antonio, Jiménez, Raúl and Yukich, Joe (2020). Localization processes for functional data analysis (submitted).

## **Examples**

```
localization Statistics\_full <- localization Statistics (example Data[,1:101], robustify = TRUE) \\ localization Statistics\_full trim\_mean[c(1, 25, 50, 75, 100)] \\ localization Statistics\_full trim\_sd[c(1, 25, 50, 75, 100)] \\
```

outlierData

Functional Gaussian processes with outliers.

## **Description**

Functional Gaussian processes with outliers.

#### Usage

outlierData

outlierLocalizationDistance 7

## **Format**

A matrix with n = 54 functions by columns and t = 200 evaluation points by row. The last 4 observations are two shape and two magnitude outliers.

#### References

Elías, Antonio, Jiménez, Raúl and Yukich, Joe (2020). Localization processes for functional data analysis (submitted).

#### **Examples**

```
matplot(outlierData, type = "l")
```

outlierLocalizationDistance

Outlier localization distances

## **Description**

Compute the localization distances of order k of the curve y0.

## Usage

```
outlierLocalizationDistance(X, localrule = 0.9, whiskerrule = 3)
```

## Arguments

Χ	matrix p by n, being n the number of functions and p the number of grid points.
localrule	Local distance rule: the method marks a curve as outlier if its k order localization distances are outliers in more than local_rulex100 percent of the k-order
	univariate boxplots. Default is 0.90 so a function must be at least an outlier in 90 percent of the k-order localization distances.
whiskerrule	Parameter for the whiskers of the univariate boxplot of the localization distances of order kth. Default value is 3.

#### Value

A list

#### References

Elías, Antonio, Jiménez, Raúl and Yukich, Joe (2020). Localization processes for functional data analysis (submitted).

## **Examples**

```
outliers <- outlierLocalizationDistance(outlierData, localrule = 0.9, whiskerrule = 3)
outliers$outliers_ld_rule</pre>
```

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